Methods for Inverse Problems: II. Nonlinear Problems and Tikhonov regularization

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Nonlinear setting

We want to solve the operator equation

$$F(x) = y \tag{1}$$

with $F : X \to Y$, X, Y Hilbert spaces, given noisy data $y^{\delta} \in Y$ s.t. $||y^{\delta} - y|| \leq \delta$. Assume that for exact data y exact solution x^{\dagger} exists and is unique.

Discuss some aspects of methods for nonlinear problems by means of the best investigated one:

Tikhonov regularization: x_{α}^{δ} minimizer of

$$J_{\alpha}(x) = \|F(x) - y^{\delta}\|^{2} + \alpha \|x - x_{0}\|^{2} = \min_{x \in D(F)} !$$
(2)

 x_0 ... initial guess of x^{\dagger} , J_{α} ... Tikhonov functional

Well-definedness and stability

F is weakly closed : \Leftrightarrow

$$\begin{array}{ll} ((\psi_n)_{n\in\mathbb{N}}\subset\mathcal{D}(F)\ \wedge\ \psi_n\rightharpoonup\psi\ \wedge\ F(\psi_n)\rightharpoonup f)\\ \implies \psi\in\mathcal{D}(F)\ \wedge\ F(\psi)=f. \end{array}$$
(3)

Theorem

Let $\alpha > 0$ and assume that F is weakly closed (3) and continuous. Then the Tikhonov functional (2) has a global minimizer.

Theorem

Let $\alpha > 0$ and assume that F is weakly closed (3) and continuous. For any sequence $y^k \to y^{\delta}$ as $k \to \infty$ the corresponding minimizers x^k_{α} of (2) (with y^k in place of y^{δ}) converge to x^{δ}_{α} .

Convergence

Theorem

Assume that F is weakly closed (3) and continuous. Let $\alpha = \overline{\alpha}(\delta)$ be chosen such that

$$\overline{\alpha}(\delta) \to 0 \quad \text{and} \quad \delta^2/\overline{\alpha}(\delta) \to 0 \qquad \text{as } \delta \to 0.$$
 (4)

If y^{δ_k} is some sequence in Y such that $||y^{\delta_k} - y|| \le \delta_k$ and $\delta_k \to 0$ as $k \to \infty$, and if $x_{\alpha_k}^{\delta_k}$ denotes a solution to (2) with $y^{\delta} = y^{\delta_k}$ and $\alpha = \alpha_k = \overline{\alpha}(\delta_k)$, then $||x_{\alpha_k}^{\delta_k} - x^{\dagger}|| \to 0$ as $k \to \infty$.

The same result holds for the discrepancy principle $\underline{\tau}\delta \leq \|F(\mathbf{x}_{\alpha}^{\delta}) - \mathbf{y}^{\delta}\| \leq \overline{\tau}\delta$ with fixed $1 < \underline{\tau} < \overline{\tau}$ in place of the a priori choice (4)

Conditions on F and convexity of the Tikhonov functional

Lemma

Let the weak nonlinearity condition [Chavent&Kunisch 1996]

 $\forall x \in D(F) \, \forall w : x + w \in D(F) : \|F''(x)[w,w]\| \le \frac{1}{R} \|F'(x)w\|^2$

for some $R > \delta$ hold. Then for all $\alpha > 0$, the Tikhonov functional is convex in a sufficiently small neighborhood of x^{\dagger} .

Proof:

$$J'_{\alpha}(x)w = 2\langle F(x) - y^{\delta}, F'(x)w \rangle + 2\alpha \langle x - x_0, w \rangle$$
$$J''_{\alpha}(x)[w, w] = 2\|F'(x)w\|^2 + 2\langle F(x) - y^{\delta}, F''(x)[w, w] \rangle + 2\alpha \|w\|^2$$

Lemma

Let the following Taylor remainder estimate $\forall \tilde{x}, x \in D(F)$

$$\|F(\tilde{x}) - F(x) - F'(x)(\tilde{x} - x)\| \le \min\{\frac{1}{R}\|F(\tilde{x}) - F(x)\|^2, \ c\|F(\tilde{x}) - F(x)\|\}$$

for some $R > 2\delta$, $c < 1 - \frac{2\delta}{R}$ hold. Then for all $\alpha > 0$, the Tikhonov functional is convex in a sufficiently small neighborhood of x^{\dagger} .

Proof:

$$\begin{aligned} \frac{1}{2} \langle J'_{\alpha}(\tilde{x}) - J'_{\alpha}(x), (\tilde{x} - x) \rangle \\ &= \langle F(\tilde{x}) - F(x), F'(x)(\tilde{x} - x) \rangle + \langle F(\tilde{x}) - y^{\delta}, (F'(\tilde{x}) - F'(x))(\tilde{x} - x) \rangle \\ &\geq \|F(\tilde{x}) - F(x)\|^2 - c\|F(\tilde{x}) - F(x)\|^2 - \frac{2}{R}\|F(x) - y^{\delta}\| \|F(\tilde{x}) - F(x)\|^2 \end{aligned}$$

since $(F'(\tilde{x}) - F'(x))(\tilde{x} - x) =$ $F(\tilde{x}) - F(x) - F'(x)(\tilde{x} - x) + F(x) - F(\tilde{x}) - F'(\tilde{x})(x - \tilde{x}).$

Convergence rates and source conditions

Theorem

Assume that F' is Lipschitz

$$\|F'[x] - F'[\tilde{x}]\| \le L \|x - \tilde{x}\|$$
(5)

for all $x, \tilde{x} \in D(F)$. Moreover, assume that the source condition

$$x^{\dagger} - x_0 = F'(x^{\dagger})^* w, \, \, \textit{with} \, \|w\| < rac{1}{L}$$

is satisfied for some $w \in Y$ and that α is chosen according to (a) the a priori parameter choice rule $\alpha = c\delta$ with some c > 0(b) the discrepancy principle $\underline{\tau}\delta \leq \|F(x_{\alpha}^{\delta}) - y^{\delta}\| \leq \overline{\tau}\delta$ with fixed $1 < \underline{\tau} < \overline{\tau}$.

Then there exists a constant C > 0 independent of δ such that

$$\|x_{\alpha}^{\delta}-x^{\dagger}\|\leq C\sqrt{\delta}, \quad \|F(x_{\alpha}^{\delta})-y\|\leq C\delta.$$

Theorem

Assume that F satisfies the tangential cone condition [Scherzer 1995]

$$\|F(\tilde{x}) - F(x) - F'(x)(\tilde{x} - x)\| \le c \|F(\tilde{x}) - F(x)\|$$
(6)

for all $x, \tilde{x} \in D(F)$. Moreover, assume that the source condition with $\mu \in (0, \frac{1}{2})$

$$x^{\dagger} - x_0 = (F'(x^{\dagger})^* F'(x^{\dagger})^{\mu} w$$

is satisfied for some $w \in X$ and that α is chosen according to (a) or (b).

Then there exists a constant C > 0 independent of δ such that

$$\|x_{lpha}^{\delta}-x^{\dagger}\|\leq C\delta^{rac{2\mu}{2\mu+1}}, \quad \|F(x_{lpha}^{\delta})-y\|\leq C\delta.$$

An example

Estimate the diffusion coefficient *a* in

$$-(a(s)u'(s))' = f(s), \quad s \in (0,1),$$

$$u(0) = 0 = u(1),$$
 (7)

where $f \in L^2$.

 $F:\mathcal{D}(F):=\{a\in H^1[0,1]: a(s)\geq \underline{a}>0\}
ightarrow L^2[0,1]$

$$\begin{array}{rcl} F(a) &=& u(a) = A(a)^{-1}f \,, \\ F'(a)h &=& A(a)^{-1}[(hu'(a))'] \,, \\ F'(a)^*w &=& -B^{-1}[u'(a)(A(a)^{-1}w)'] \,, \end{array}$$

•
$$A(a): H^2[0,1] \cap H^1_0[0,1] \to L^2[0,1],$$

 $A(a)u := -(au')'$
• $B^{-1} \dots$ adjoint of the embedding $H^1[0,1] \to L^2[0,1].$
 $B: \{\psi \in H^2[0,1]: \psi'(0) = \psi'(1) = 0\} \to L^2[0,1],$
 $B\psi := -\psi'' + \psi.$

$$\langle F(\tilde{a}) - F(a) - F'(a)(\tilde{a} - a), w \rangle_{L^{2}} = \langle (\tilde{u} - u) - A(a)^{-1}[((\tilde{a} - a)u')'], w \rangle_{L^{2}} = \langle A(a)(\tilde{u} - u) - ((\tilde{a} - a)u')', A(a)^{-1}w \rangle_{L^{2}} = \langle ((\tilde{a} - a)(\tilde{u}' - u'))', A(a)^{-1}w \rangle_{L^{2}} = -\langle (\tilde{a} - a)(\tilde{u} - u)', (A(a)^{-1}w)' \rangle_{L^{2}} = \langle F(\tilde{a}) - F(a), ((\tilde{a} - a)(A(a)^{-1}w)')' \rangle_{L^{2}}.$$

$$\begin{split} \| ((\tilde{a} - a)(A(a)^{-1}w)')'\|_{L^{2}} &= \| (\frac{\tilde{a} - a}{a}) a(A(a)^{-1}w)')'\|_{L^{2}} \\ &= \| (\frac{\tilde{a} - a}{a})' a(A(a)^{-1}w)' + \frac{\tilde{a} - a}{a} (a(A(a)^{-1}w')'\|_{L^{2}} \\ &\leq \frac{1}{a} \Big(\| (\tilde{a} - a)'\|_{L^{2}} + \frac{1}{a^{2}} \| a'\|_{L^{2}} \| \tilde{a} - a)\|_{L^{\infty}} \Big) \underbrace{\| a(A(a)^{-1}w)'\|_{L^{\infty}}}_{\leq \bar{a}\hat{C} \| (A(a)^{-1}w)'\|_{H^{1}} \leq \bar{a}\hat{C} \| A(a)^{-1}\|_{L^{2} \to H^{2}} \| w \|_{L^{2}} \Big) \\ &\leq \bar{a}\hat{C} \| (A(a)^{-1}w)'\|_{H^{1}} \leq \bar{a}\hat{C} \| A(a)^{-1}\|_{L^{2} \to H^{2}} \| w \|_{L^{2}} \end{split}$$

$$\Rightarrow \|F(\tilde{a}) - F(a) - F'(a)(\tilde{a} - a)\|_{L^2} \\ \leq C \|F(\tilde{a}) - F(a)\|_{L^2} \|\tilde{a} - a\|_{H^1}.$$

Literature:

- stability and convergence: [Seidman&Vogel 1989]
- convergence rates [Engl&Kunisch&Neubauer 1989] [Neubauer 1999] [Hofmann&Scherzer 1998]
- analysis in Banach space: [Burger&Osher 2004], [Hofmann&Kaltenbacher&Pöschl&Scherzer 2007]
 [Hein&Hofmann&Kindermann&Neubauer&Tautenhahn 2009]

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